

statistics

February 1, 2021

```
[62]: import pandas as pd
import matplotlib.pyplot as plt
import matplotlib.dates as mdates
from matplotlib.gridspec import GridSpec
%matplotlib inline

import datetime
import geopandas as gpd

import numpy as np
import seaborn as sns
```

```
[63]: import warnings
warnings.filterwarnings("ignore")

pd.set_option("display.float_format", lambda x: "%.2f" % x)
```

```
[64]: #importing data using pandas

df_covid = pd.read_csv("covidde/covid_de.csv")
df_dmo = pd.read_csv("covidde/demographics_de.csv")

#reading .shp (shapeFile) file using geopandas. geopandas extends the datatypes_
↳used by pandas
df_mp = gpd.read_file("covidde/de_state.shp")
df_mp_country = gpd.read_file("covidde/de_county.shp")
```

```
[65]: df_covid
```

```
[65]:
```

	state	county	age_group	gender	date	\
0	Baden-Wuerttemberg	LK Alb-Donau-Kreis	00-04	F	2020-03-27	
1	Baden-Wuerttemberg	LK Alb-Donau-Kreis	00-04	F	2020-03-28	
2	Baden-Wuerttemberg	LK Alb-Donau-Kreis	00-04	F	2020-04-03	
3	Baden-Wuerttemberg	LK Alb-Donau-Kreis	00-04	F	2020-10-18	
4	Baden-Wuerttemberg	LK Alb-Donau-Kreis	00-04	F	2020-10-22	
...	

416309	Thueringen	SK Weimar	80-99	M	2020-12-26
416310	Thueringen	SK Weimar	80-99	M	2020-12-28
416311	Thueringen	SK Weimar	80-99	M	2020-12-30
416312	Thueringen	SK Weimar	80-99	M	2020-12-31
416313	Thueringen	SK Weimar	NaN	F	2020-12-31

	cases	deaths	recovered
0	1	0	1
1	1	0	1
2	1	0	1
3	1	0	1
4	1	0	1
...
416309	4	1	0
416310	1	0	0
416311	2	0	0
416312	1	0	0
416313	1	0	0

[416314 rows x 8 columns]

[66]: *#changing the name of state column value*

```
df_covid["state"] = df_covid["state"].replace("Baden-Wuerttemberg",
↪ "Baden-Württemberg")
df_covid["state"] = df_covid["state"].replace("Thueringen", "Thüringen")
```

[67]: df_covid

[67]:

	state	county	age_group	gender	date \
0	Baden-Württemberg	LK Alb-Donau-Kreis	00-04	F	2020-03-27
1	Baden-Württemberg	LK Alb-Donau-Kreis	00-04	F	2020-03-28
2	Baden-Württemberg	LK Alb-Donau-Kreis	00-04	F	2020-04-03
3	Baden-Württemberg	LK Alb-Donau-Kreis	00-04	F	2020-10-18
4	Baden-Württemberg	LK Alb-Donau-Kreis	00-04	F	2020-10-22
...
416309	Thüringen	SK Weimar	80-99	M	2020-12-26
416310	Thüringen	SK Weimar	80-99	M	2020-12-28
416311	Thüringen	SK Weimar	80-99	M	2020-12-30
416312	Thüringen	SK Weimar	80-99	M	2020-12-31
416313	Thüringen	SK Weimar	NaN	F	2020-12-31

	cases	deaths	recovered
0	1	0	1
1	1	0	1
2	1	0	1
3	1	0	1

4	1	0	1
...
416309	4	1	0
416310	1	0	0
416311	2	0	0
416312	1	0	0
416313	1	0	0

[416314 rows x 8 columns]

```
[68]: #formatting date value
df_covid["date"] = pd.to_datetime(df_covid["date"])
```

```
[69]: #changing the name of state column value

df_dmo["state"] = df_dmo["state"].replace("Baden-Wuerttemberg",
↪ "Baden-Württemberg")
df_dmo["state"] = df_dmo["state"].replace("Thueringen", "Thüringen")
```

```
[70]: #changing the value of gender column of df_dmp dataframe

df_dmo["gender"] = np.where(df_dmo["gender"] == "female", "F", "M" )
```

```
[71]: df_dmo
```

```
[71]:
```

	state	gender	age_group	population
0	Baden-Württemberg	F	00-04	261674
1	Baden-Württemberg	F	05-14	490822
2	Baden-Württemberg	F	15-34	1293488
3	Baden-Württemberg	F	35-59	1919649
4	Baden-Württemberg	F	60-79	1182736
..
187	Thüringen	M	05-14	92545
188	Thüringen	M	15-34	214553
189	Thüringen	M	35-59	384822
190	Thüringen	M	60-79	264189
191	Thüringen	M	80-99	57340

[192 rows x 4 columns]

```
[72]: #finding 'NA' values and displaying them"

df_covid[(df_covid["gender"].isnull()) | (df_covid["age_group"].isnull())]
```

```
[72]:
```

	state	county	age_group	gender	date \
200	Baden-Württemberg	LK Alb-Donau-Kreis	05-14	NaN	2020-10-30
201	Baden-Württemberg	LK Alb-Donau-Kreis	05-14	NaN	2020-11-19

517	Baden-Württemberg	LK Alb-Donau-Kreis	15-34	NaN	2020-10-28
518	Baden-Württemberg	LK Alb-Donau-Kreis	15-34	NaN	2020-10-30
519	Baden-Württemberg	LK Alb-Donau-Kreis	15-34	NaN	2020-11-01
...
415484	Thüringen	SK Jena	NaN	F	2020-12-29
415485	Thüringen	SK Jena	NaN	M	2020-12-28
415486	Thüringen	SK Jena	NaN	M	2020-12-29
415487	Thüringen	SK Jena	NaN	M	2020-12-30
416313	Thüringen	SK Weimar	NaN	F	2020-12-31

	cases	deaths	recovered
200	1	0	1
201	1	0	1
517	1	0	1
518	2	0	2
519	1	0	1
...
415484	1	0	0
415485	1	0	0
415486	1	0	0
415487	2	0	0
416313	1	0	0

[11402 rows x 8 columns]

```
[73]: df_dmo[(df_dmo["gender"].isnull()) | (df_dmo["age_group"].isnull())]
```

```
[73]: Empty DataFrame
Columns: [state, gender, age_group, population]
Index: []
```

#This data set has no “NaN” values

```
[74]: #getting the sum of the 'NA' values

df_covid[(df_covid["gender"].isnull()) | (df_covid["age_group"].isnull())].sum()
```

```
[74]: state      Baden-WürttembergBaden-WürttembergBaden-Württe...
county      LK Alb-Donau-KreisLK Alb-Donau-KreisLK Alb-Don...
cases                                             17160
deaths                                             65
recovered                                         12189
dtype: object
```

```
[75]: #filling age_group "NA" values with most frequent values

group = df_covid.age_group.value_counts().idxmax()
```

```
df_covid.age_group.fillna(group, inplace = True)
```

```
[76]: #filling missing values of gender column (half with 'M' and other half with 'F')

gender = df_covid.gender.isna()
gen = df_covid.gender.loc[gender].sample(frac = 0.5).index #generating random
↳rows
df_covid.loc[gen, 'gender'] = 'M'
df_covid.gender.fillna('F', inplace = True)
```

##Now the 'NA' values is filled up. We can Check it by calling the isnull() function

```
[77]: df_covid[(df_covid["gender"].isnull()) | (df_covid["age_group"].isnull())]
```

```
[77]: Empty DataFrame
Columns: [state, county, age_group, gender, date, cases, deaths, recovered]
Index: []
```

```
[78]: #getting mean,max, total values from dataset

df_covid.describe()
```

```
[78]:
```

	cases	deaths	recovered
count	416314.00	416314.00	416314.00
mean	4.22	0.08	3.29
std	6.71	0.38	5.62
min	0.00	0.00	-2.00
25%	1.00	0.00	1.00
50%	2.00	0.00	1.00
75%	4.00	0.00	3.00
max	206.00	13.00	206.00

##We will start analyzing our data. For that purpose we will use different kind of charts, plots from Matplotlib, Seaborn and GeoPandas library. Let's see what we can do with the mentioned library

```
[79]: #plotting the covid cases according to the month

daily_cases = df_covid.groupby("date").sum()

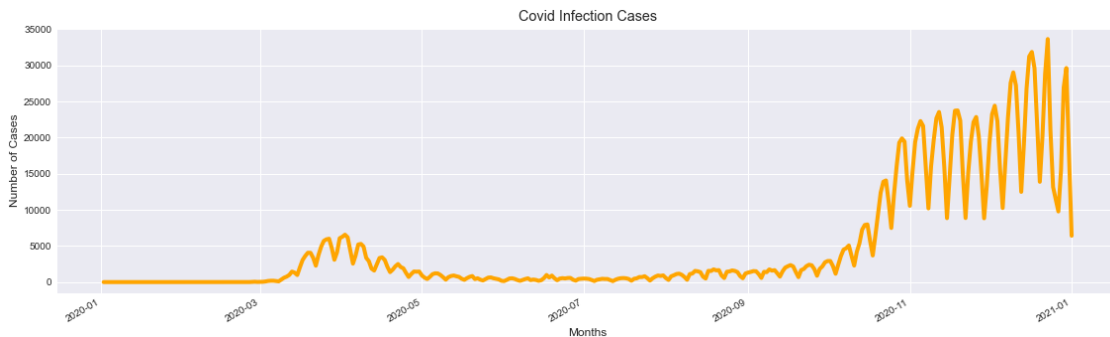
sns.set_style("darkgrid")
#sns.color_palette("cool")

plt.figure(figsize=(16,5))
plt.title("Covid Infection Cases")
sns.lineplot(data = daily_cases["cases"], color = "orange")
```

```
plt.ylabel("Number of Cases")
plt.xlabel("Months")

plt.style.use("fivethirtyeight")

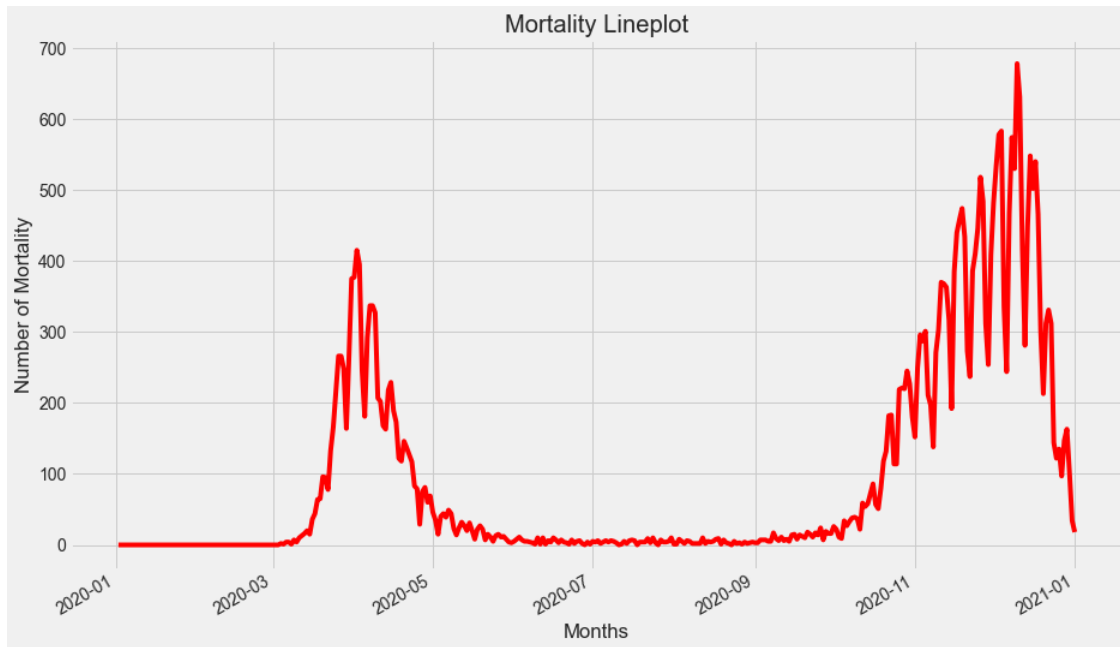
plt.gcf().autofmt_xdate()
plt.tight_layout()
plt.show()
```



##From the lineplot of covid_cases we can say that, infection rate were almost 0 between January to March. Between March and May the number of infected people increases compare to the first 3 months. After that, till October the amount of cases were under control before significantly increasing during the period of November to January. This period is called the second wave of Covid.

```
[80]: #daily mortality

plt.figure(figsize=(14,9))
plt.title("Mortality Lineplot")
sns.lineplot(data = daily_cases["deaths"], color = "red")
plt.ylabel("Number of Mortality")
plt.xlabel("Months")
plt.gcf().autofmt_xdate()
plt.show()
```



##The death cases followed the same graph as covid cases but the numbers were significantly low.The Death cases reached highest level of the graph during December and January.

```
[81]: #which age groups are most affected by the virus

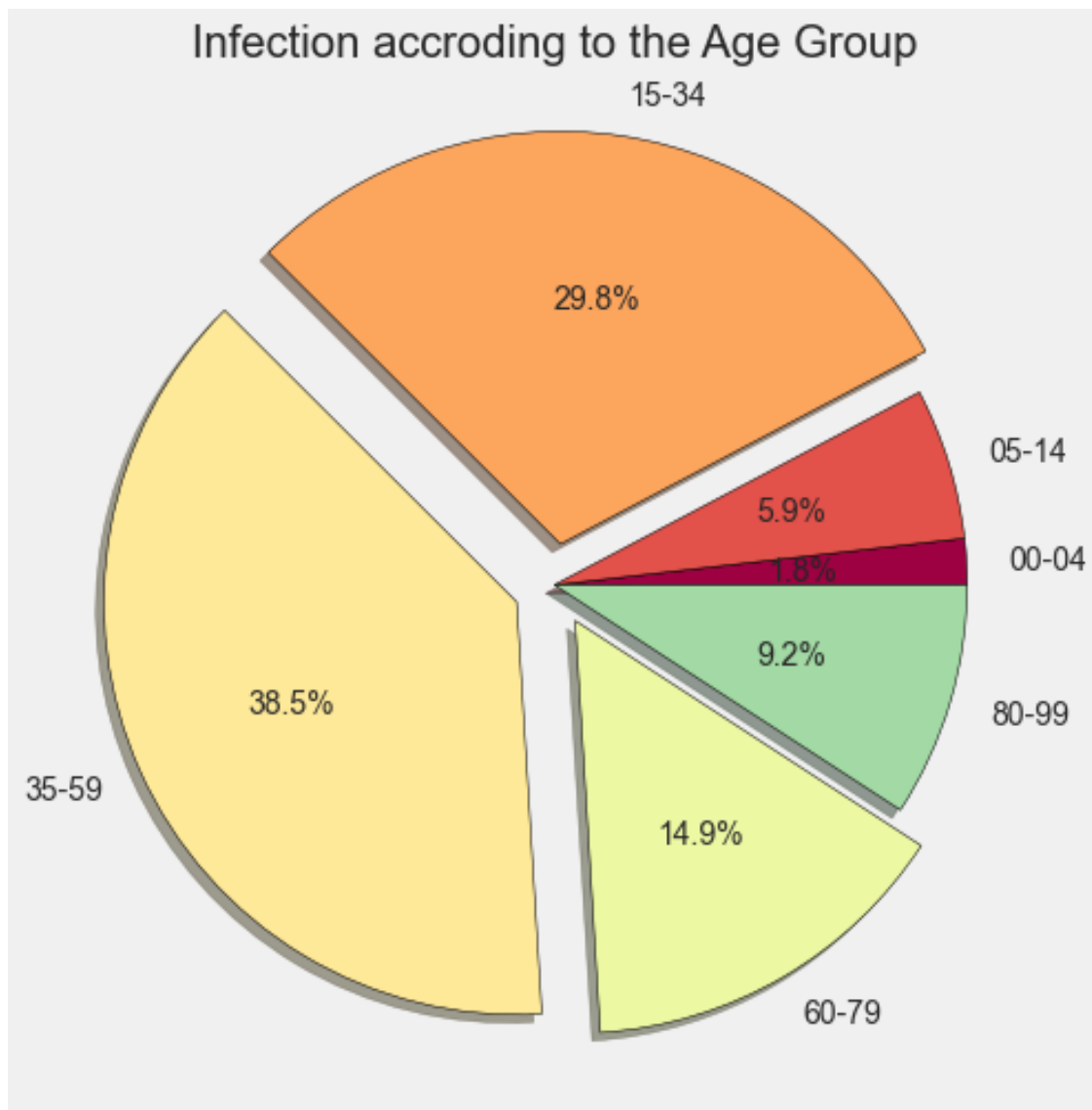
plt.style.use("fivethirtyeight")
plt.tight_layout()

by_age = df_covid.groupby("age_group").sum()
by_age.sort_values("cases",ascending = False)

cmap = plt.get_cmap("Spectral")
colors = [cmap(i) for i in np.linspace(0, 1, 8)]
explode = [0,0,0.1,0.1,0.1,0]

plt.figure(figsize = (10,8))
plt.title("Infection accroding to the Age Group")
cases_pie = plt.pie(by_age.cases, labels = by_age.index, autopct = "%1.
    ↳1f%",shadow = True , colors = colors
                    ,wedgeprops={'edgecolor':'black'},explode = explode)
```

<Figure size 432x288 with 0 Axes>



##From the pie chart we can tell that people of age group of 35-59 are the most infected age group. But covid was not nice to the elderly and younger age groups.

```
[82]: #How the number of deaths is distributed over different age groups of both
      ↪ genders

covid = df_covid.groupby(['age_group', 'gender'], as_index = False).sum()

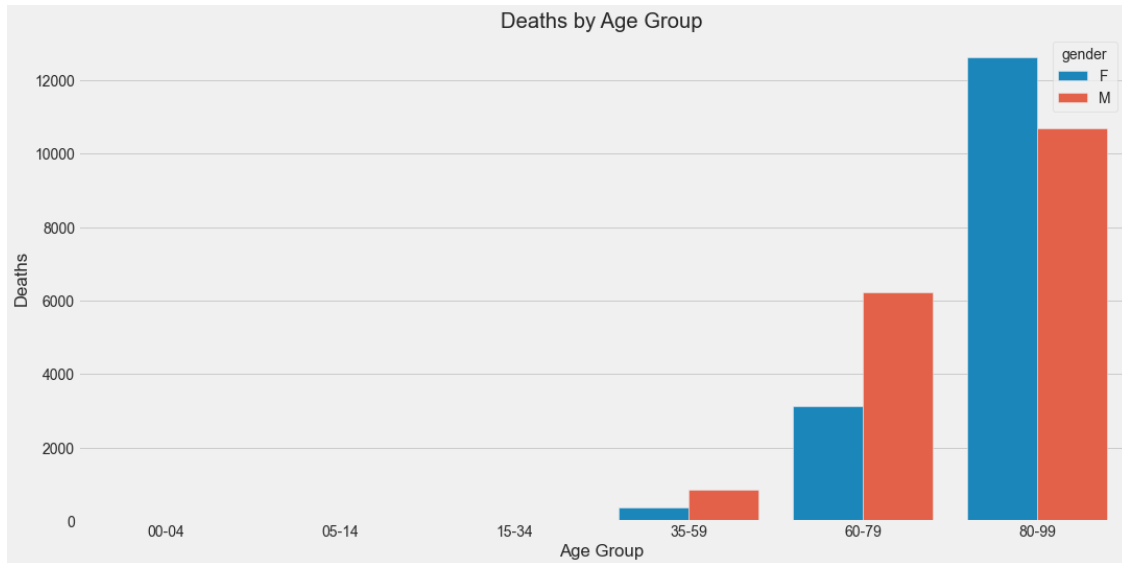
plt.figure(figsize=(16,8))
sns.barplot(y = covid.deaths, x = covid.age_group, hue = covid.gender, data =
      ↪ covid)

plt.style.use("fivethirtyeight")
```



```
plt.title('Deaths by Age Group')
plt.xlabel('Age Group')
plt.ylabel('Deaths')
```

[82]: Text(0, 0.5, 'Deaths')



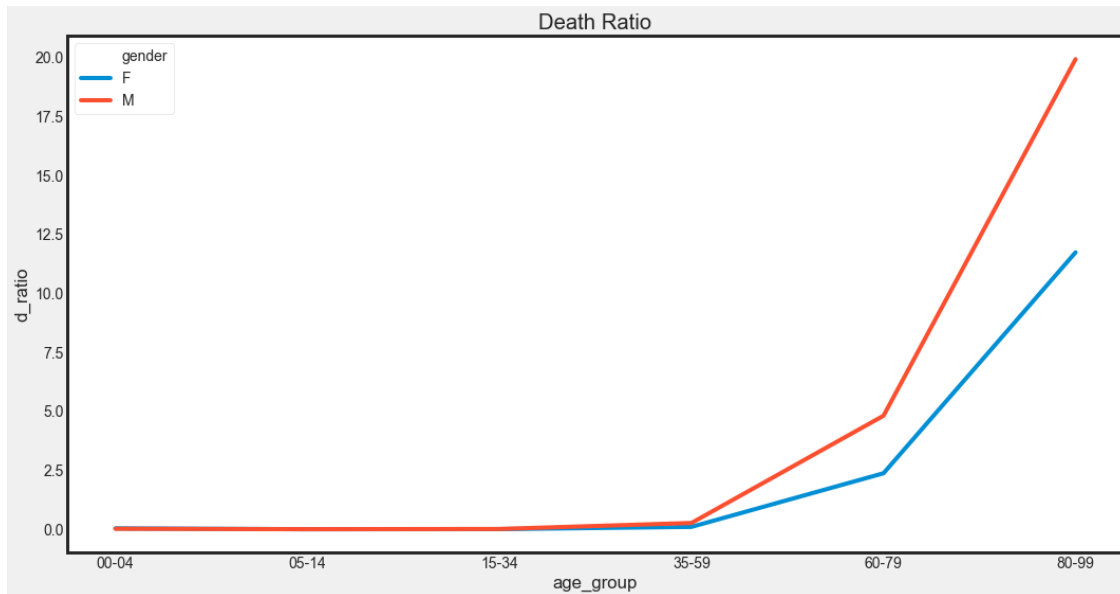
##This barplot indicates that females in the 80-99 age group died more by the virus than males. The ratio is half and totally opposite in the age group of 60-79.

[83]: *#death ratio by gender*

```
gender = df_covid.groupby(by=["age_group", "gender"]).sum().reset_index()
gender["d_ratio"] = 100 * gender["deaths"] / gender["cases"]

plt.figure(figsize=(15,8))
sns.set_style("ticks")
sns.lineplot(data = gender,x="age_group",y="d_ratio",hue = "gender")
plt.title("Death Ratio")

plt.tight_layout()
plt.show()
```



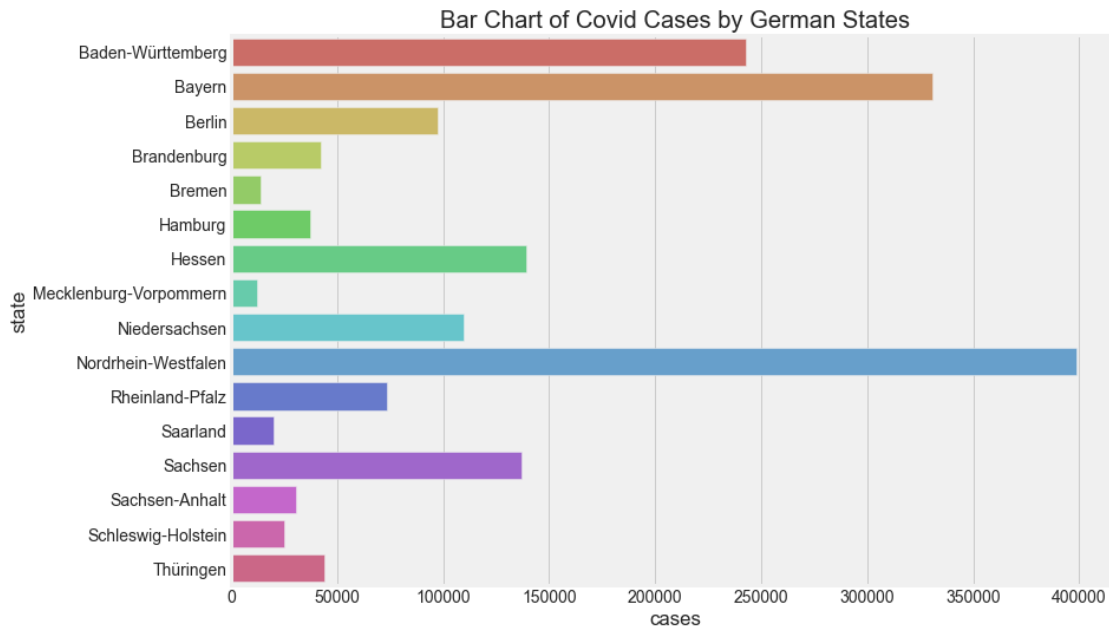
#This lineplot illustrates that death ratio is higher in elderly male age-group compare to female age groups.

```
[84]: #states comparison

by_state = df_covid.groupby("state").sum()
by_state.sort_values("cases",ascending = True)
#width = 0.25

plt.figure(figsize=(12,8))
plt.style.use("fivethirtyeight")
plt.tight_layout()

plt.title("Bar Chart of Covid Cases by German States")
#plt.xlabel("Number of Total Cases")
#plt.ylabel("Name of State")
sns.barplot(by_state.cases, by_state.index, palette= 'hls')
#plt.legend()
plt.show()
```



We can see from the barplot above that, Nordrhein-Westfalen has got the most coronavirus cases. The least infected state is Hessen and Bremen. Bayern and Baden-Württemberg are also in the track of most infection

```
[85]: df_mp.head()
```

```
[85]:   ADE  RS   RS_0      GEN \
0    2  02  020000000000      Hamburg
1    2  03  030000000000  Niedersachsen
2    2  04  040000000000      Bremen
3    2  05  050000000000  Nordrhein-Westfalen
4    2  06  060000000000      Hessen

                                geometry
0  MULTIPOLYGON (((3578695.661 5955304.456, 35781...
1  MULTIPOLYGON (((3354775.046 5942939.764, 33546...
2  MULTIPOLYGON (((3468658.496 5898364.974, 34702...
3  POLYGON ((3477450.781 5820982.368, 3479895.578...
4  POLYGON ((3535084.230 5721608.644, 3535279.888...
```

```
[86]: #dropping unnecessary columns from dataset
```

```
df_mp = df_mp.drop(columns = ["ADE", "RS", "RS_0"])
```

```
[87]: df_mp.head()
```

```
[87]:
```

	GEN	geometry
0	Hamburg	MULTIPOLYGON (((3578695.661 5955304.456, 35781...
1	Niedersachsen	MULTIPOLYGON (((3354775.046 5942939.764, 33546...
2	Bremen	MULTIPOLYGON (((3468658.496 5898364.974, 34702...
3	Nordrhein-Westfalen	POLYGON ((3477450.781 5820982.368, 3479895.578...
4	Hessen	POLYGON ((3535084.230 5721608.644, 3535279.888...

```
[88]: state_covid = df_covid.groupby(by = 'state',as_index=False).sum()
state_demo = df_dmo[['state', 'population']].groupby(by='state',as_index =
↳False).sum()
df_state = df_mp.merge(state_covid, how = "left", left_on = "GEN", right_on =
↳"state") #merging 2 dataset together and alligning the columns
df_state = df_state.merge(state_demo, how = "left", left_on = "GEN", right_on =
↳"state")
df_state = df_state.drop(columns = ["state_x", "state_y"]) #dropping state
↳column to avoid multiple same value column existence

df_state["case_ratio"] = df_state["cases"] * (1000 / df_state ["population"])
↳#getting case ratio per 1000 in a new column
df_state["d_ratio"] = df_state["deaths"] * (1000 / df_state ["population"])
↳#getting death case ratio per 1000 in a new column
df_state["d_case_ratio"] = 100 * df_state["deaths"] / df_state["cases"]
↳#getting death-case ratio in a new column

df_state.set_index("GEN", inplace = True)
#state_covid
#state_demo
df_state.head()
```

```
[88]:
```

	GEN	geometry \
	Hamburg	MULTIPOLYGON (((3578695.661 5955304.456, 35781...
	Niedersachsen	MULTIPOLYGON (((3354775.046 5942939.764, 33546...
	Bremen	MULTIPOLYGON (((3468658.496 5898364.974, 34702...
	Nordrhein-Westfalen	POLYGON ((3477450.781 5820982.368, 3479895.578...
	Hessen	POLYGON ((3535084.230 5721608.644, 3535279.888...

	cases	deaths	recovered	population	case_ratio \
GEN					
Hamburg	37286	658	27236	1841179	20.25
Niedersachsen	109797	2016	92551	7982448	13.75
Bremen	13700	201	11380	682986	20.06
Nordrhein-Westfalen	398661	6701	326307	17932651	22.23
Hessen	139349	2917	109223	6265809	22.24

	d_ratio	d_case_ratio
GEN		

Hamburg	0.36	1.76
Niedersachsen	0.25	1.84
Bremen	0.29	1.47
Nordrhein-Westfalen	0.37	1.68
Hessen	0.47	2.09

#We have merged 3 different dataset in a single dataset. We have added some new columns in our new dataset. It will help us to visualize our geographical overview. case_ratio, death ratio and death case ratio column will be used in geographical graphs.

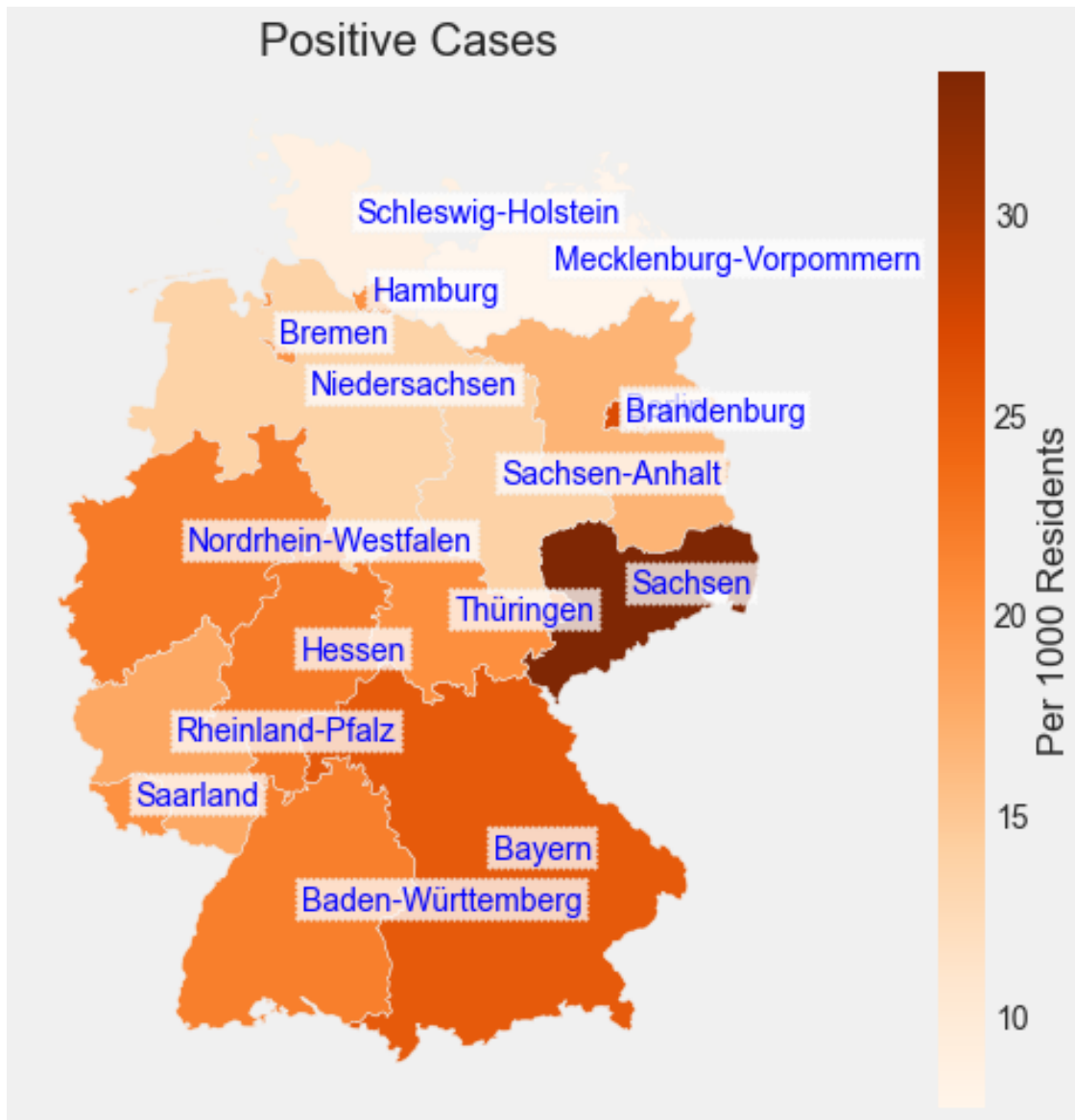
```
[89]: fig, (ax) = plt.subplots(1, figsize=(20,8), edgecolor = 'black')

df_state.plot(column = "case_ratio", legend = True, legend_kwds = {"label": "Per 1000 Residents",
    "orientation" : "vertical"}, cmap = "Oranges", ax = ax)

for i, geo in df_state.centroid.iteritems():
    ax.annotate(s=i, xy=[geo.x , geo.y], color = "blue", bbox = dict(boxstyle = "sawtooth, pad =0.2" , fc = 'white', alpha = 0.75))

    ax.set_title("Positive Cases")

ax.axes.get_xaxis().set_visible(False)
ax.axes.get_yaxis().set_visible(False)
```



[]:

```
[90]: fig, (ax1) = plt.subplots(1, figsize=(20,8), edgecolor = 'black')

df_state.plot(column="d_case_ratio", legend = True, legend_kwds={"label": "In_
    ↳Percentage (%)",

    ↳"orientation": "vertical"}, cmap = "Oranges", ax = ax1)

for i, geo in df_state.centroid.iteritems():
```

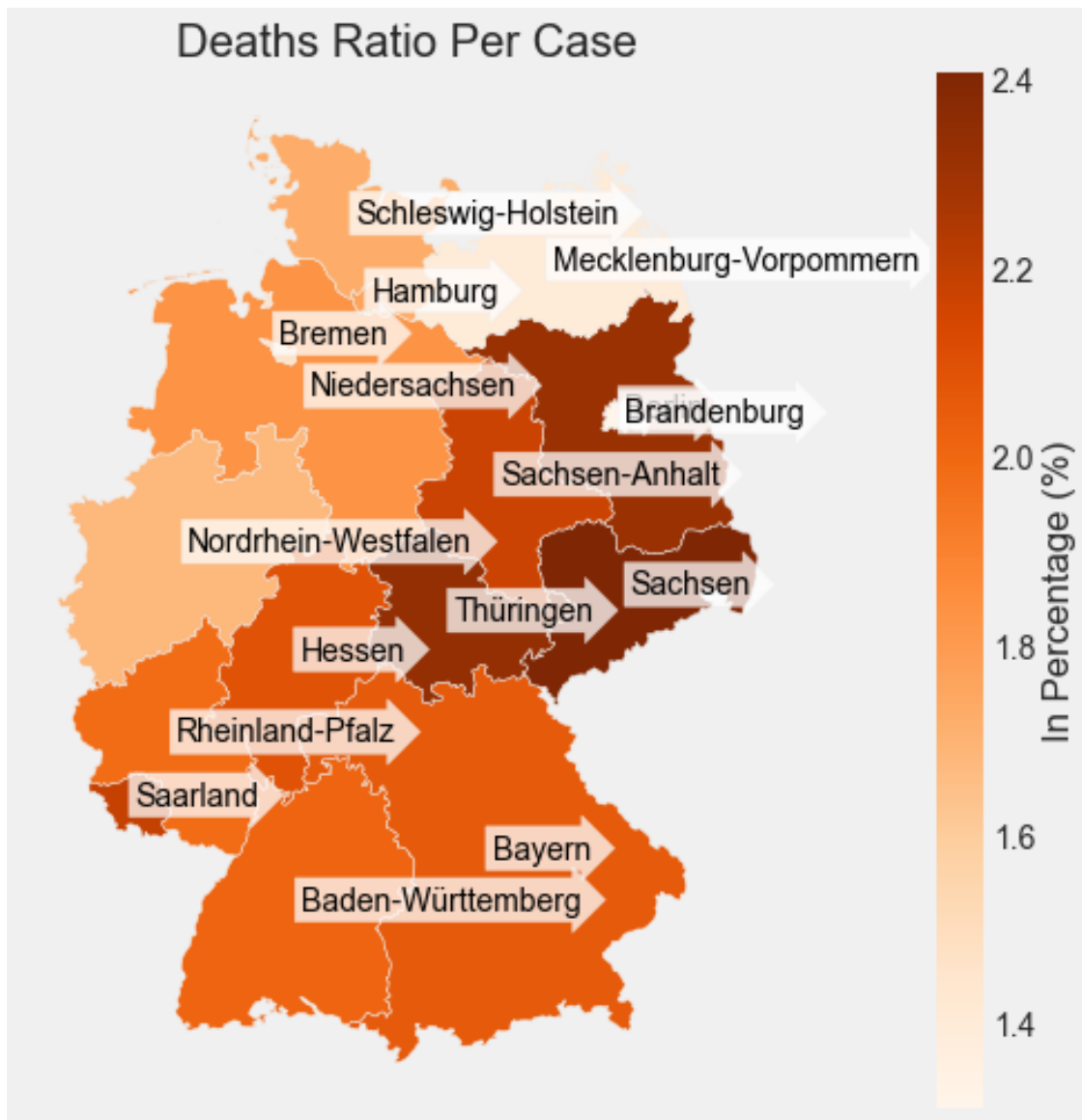
```

ax1.annotate(s=i, xy=[geo.x, geo.y], color = "black", bbox = dict(boxstyle=
↳ 'rarrow, pad =0.2', fc = 'white', alpha = 0.75))
ax1.set_title("Deaths Ratio Per Case")

ax1.axes.get_xaxis().set_visible(False)
ax1.axes.get_yaxis().set_visible(False)

plt.show()

```



#Using the geomatic value of our dataset we have tried to visualize all the states of Germany according to Positive cases and Death ratio per case. They have been calculated above. Because of using multiple dataset for displaying our geomatic data there might be a slight change between

state comparison bar graph and geo graph. The data displayed above might seem diverse.

1 THANK YOU

[]: